

Computational study on biomass fast pyrolysis:

# Hydrodynamic effects on the performance of a laboratory-scale fluidized bed reactor

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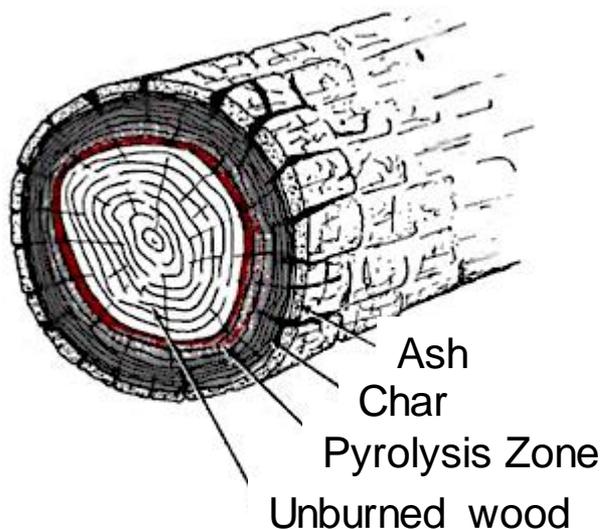
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National Energy Technology Laboratory's (NETL)  
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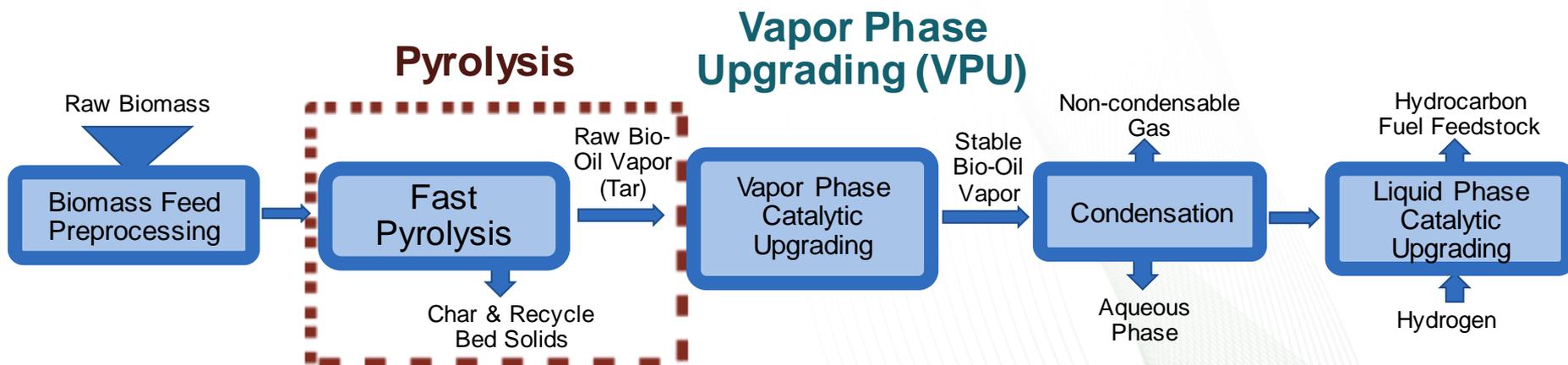


# Thermochemical conversion of biomass via fast pyrolysis



Source: Heat-Its Role in Wildland Fire by CiveM. Countyman

- High yield and composition of raw oil are key, so commercial risk and economics depend on accurate performance predictions.
- Most available basic lab data are from bubbling fluidized bed reactors (FBRs).
- Good physics-based models are necessary for interpreting and scaling up lab experiments.



# How do bubbling-bed hydrodynamics affect raw oil yield & composition?

Hydrodynamics directly impact:

1. Particle residence time
2. Gas residence time
3. Particle heating rate
4. Particle attrition/fragmentation
5. Particle and ash elutriation
6. Particle segregation

All the above significantly impact raw oil yield and composition.

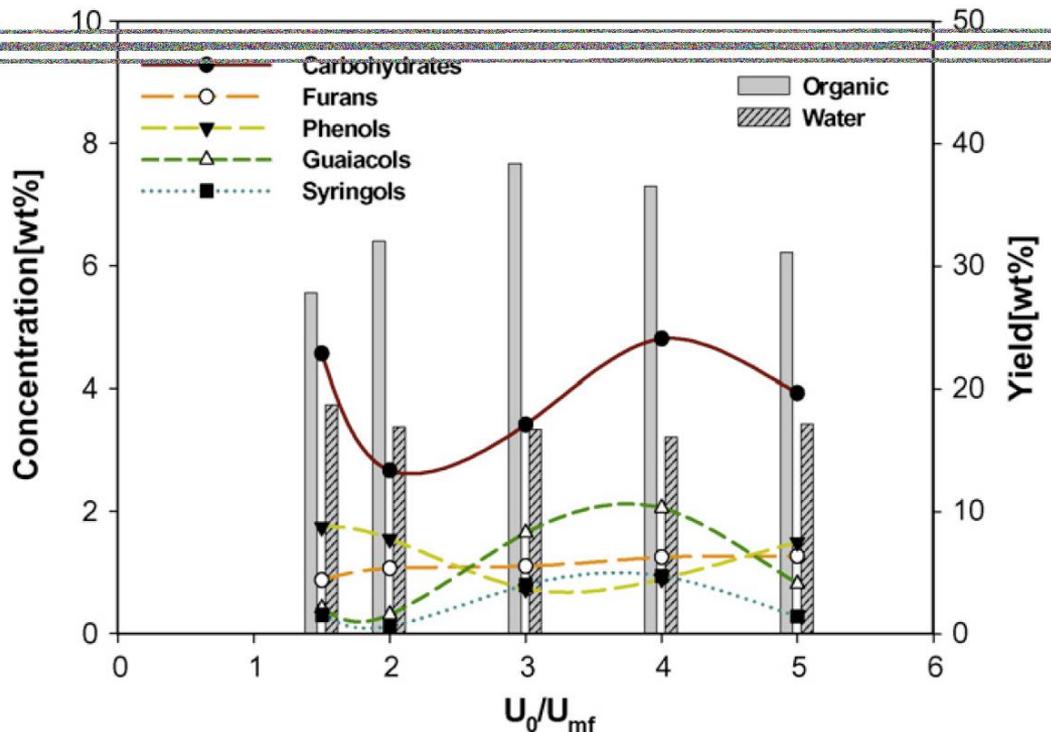


Figure: S.-H. Lee, M.-S. Eom, K.-S. Yoo, N.-C. Kim, J.-K. Jeon, Y.-K. Park, B.-H. Song, S.-H. Lee, The yields and composition of bio-oil produced from *quercus acutissima* in a bubbling fluidized bed pyrolyzer, J. Anal. Appl. Pyrolysis 83 (2008) 110-114. <http://dx.doi.org/10.1016/j.jaap.2008.06.006>

## MFiX simulations of FBR pyrolysis

### Two-Fluid Model

- Version and assumptions:

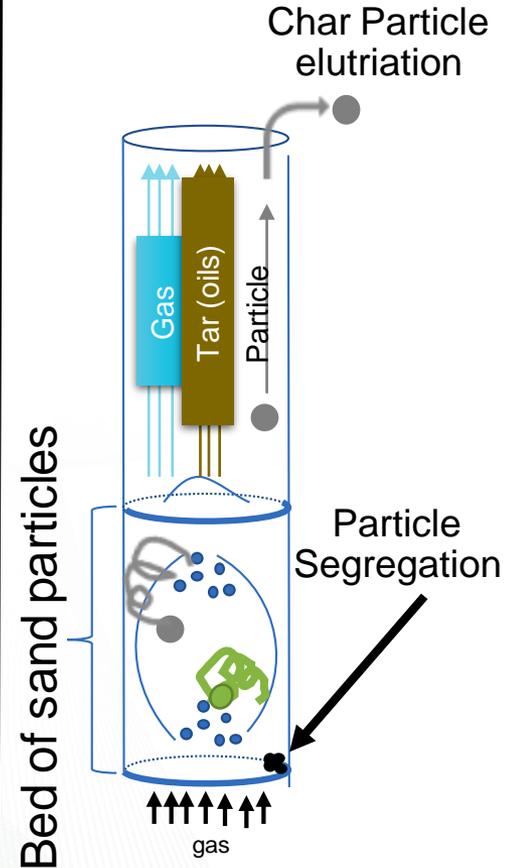
- Eulerian-Eulerian (Two-Fluid Model)
- Syamlal-O'Brien drag model
- Kinetic theory of granular flow
  - Schaeffer frictional stress tensor formulation
  - Sigmoidal stress blending function
- Modified SIMPLE integration with variable time stepping
- Jackson and Johnson partial-slip wall boundary condition
- 3D cylindrical
- Constant biomass density (char)
- Liden reduced kinetics for biomass
- DLSODA ODEPACK chemistry solver
  - First-order irreversible Arrhenius rates
  - Liden 1988 biomass pyrolysis kinetics



$$\frac{dm_i}{dt} = -mk_i$$

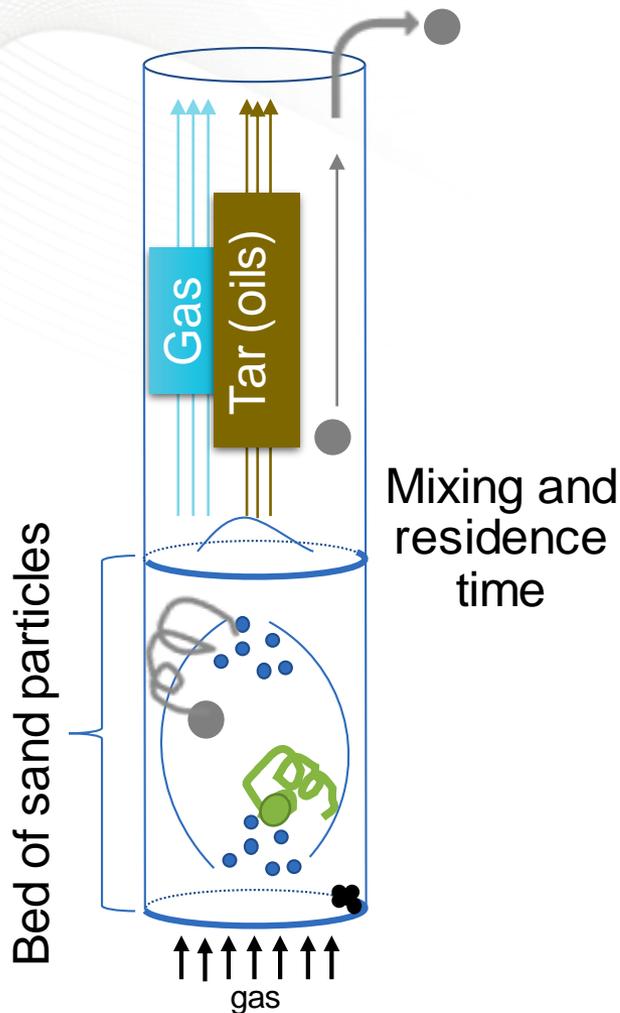
$$k_i = A_i \exp(-E_i / RT)$$

### Pyrolysis reactor physics



Ramirez, E., Finney, C. E. A., Pannala, S., Daw, C. S., Halow, J., & Xiong, Q. (2017). Computational study of the bubbling-to-slugging transition in a laboratory-scale fluidized bed. *Chemical Engineering Journal*, 308, 544-556. doi:https://doi.org/10.1016/j.cej.2016.08.113

## Interpret MFIX Results with Low-Order Models



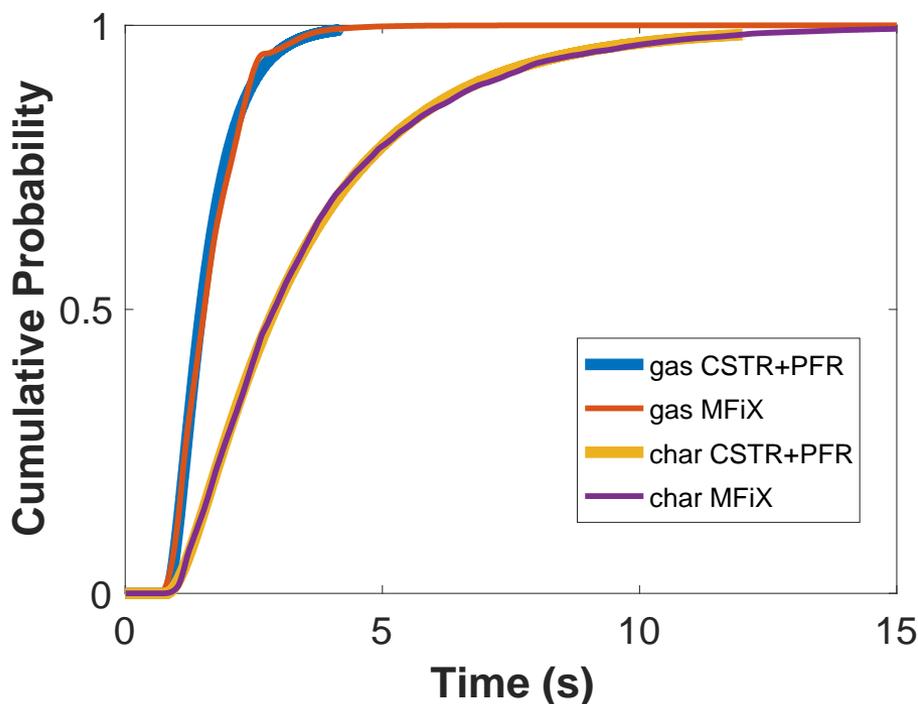
Use simplified reactor models to 'compress' essential hydrodynamic information from MFIX and combine it with pyrolysis chemistry

- Quantify impact of bubbles and bed solids circulation on biomass solids and pyrolysis vapor residence time distributions (RTDs)
- Identify major reaction/mixing zones needed to understand/approximate performance trends
- Relate solids and gas RTDs to predict trends for how biomass particle properties and reaction chemistry impact overall yields
- Utilize low-order models for rapid studies of operating/design parameter sweeps

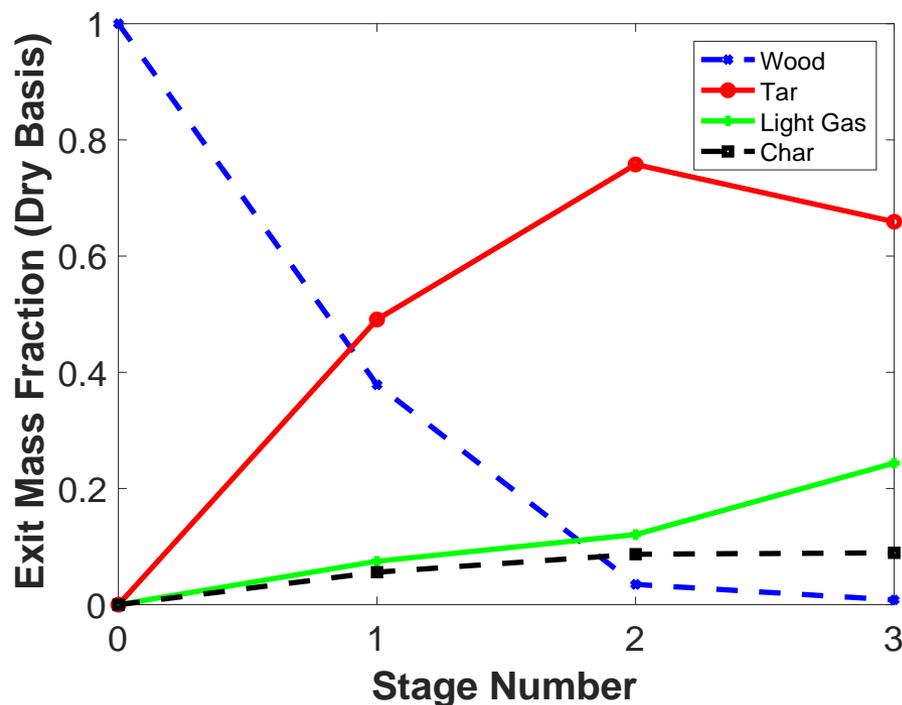
## Approach (3)

# Fast pyrolysis model using low-order chemistry + MFiX-based hydrodynamics ('Hybrid' modeling approach)

Step 1. Use MFiX gas and biomass RTDs to create zone reactor model approximation



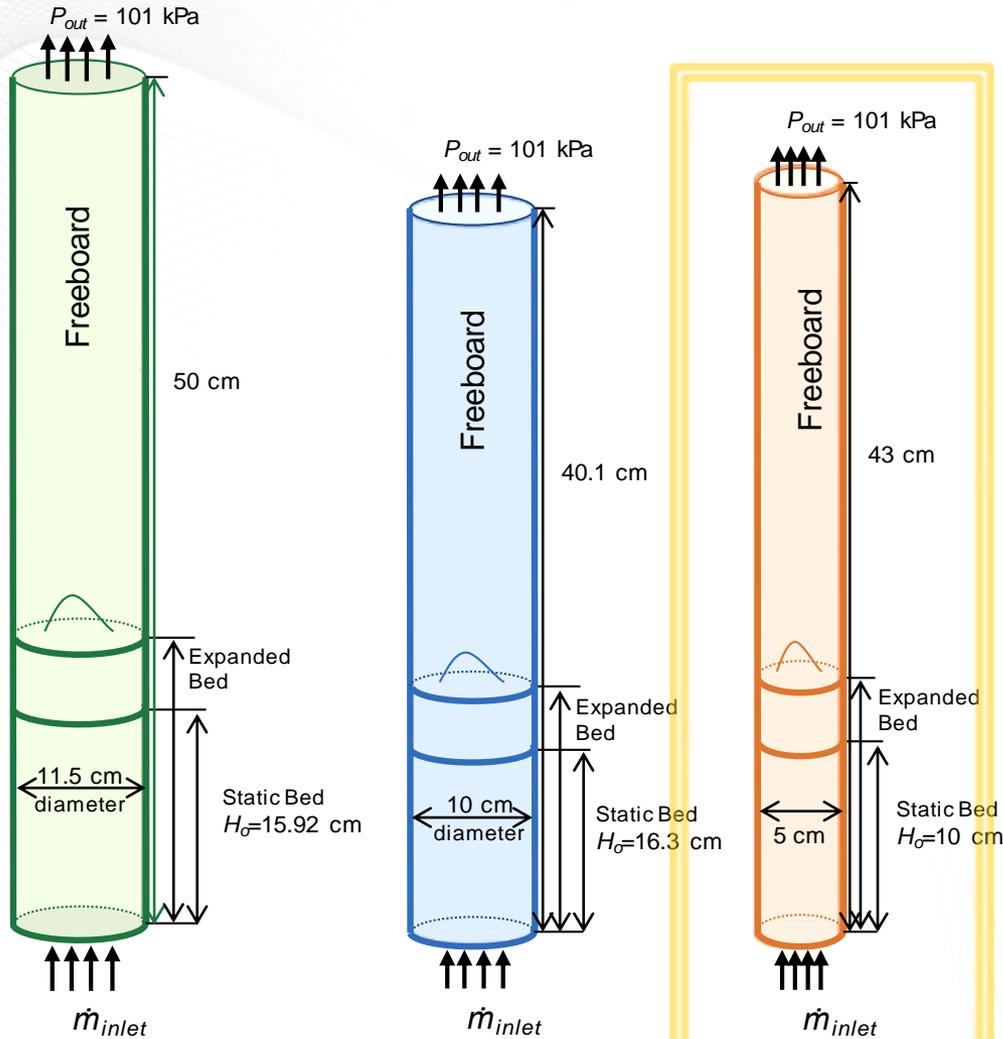
Step 2. Use zone model + Liden kinetics to predict yields



E. Ramirez, Tingwen Li, Mehrdad Shahn timer, C. Stuart Daw, Computational study on biomass fast pyrolysis: Hydrodynamic effects on the performance of a laboratory-scale fluidized bed reactor. [Manuscript in preparation.](#)

## Approach (4)

# Compare MFiX predictions with target reactor data



**Mixing biomass char**  
Park and Choi 2013

**RTD study**  
Berruti 1988

**NREL Pyrolysis Experiment**

### Key steps:

- Simulate expected particle and gas RTDs with MFiX including segregation and elutriation

### Questions:

- Are MFiX mixing patterns consistent with the literature?
- Can existing FB correlations capture MFiX predicted RTD trends?
- When chemistry is added, do predicted bio-oil yields agree with experiments?
- Are MFiX improvements needed?

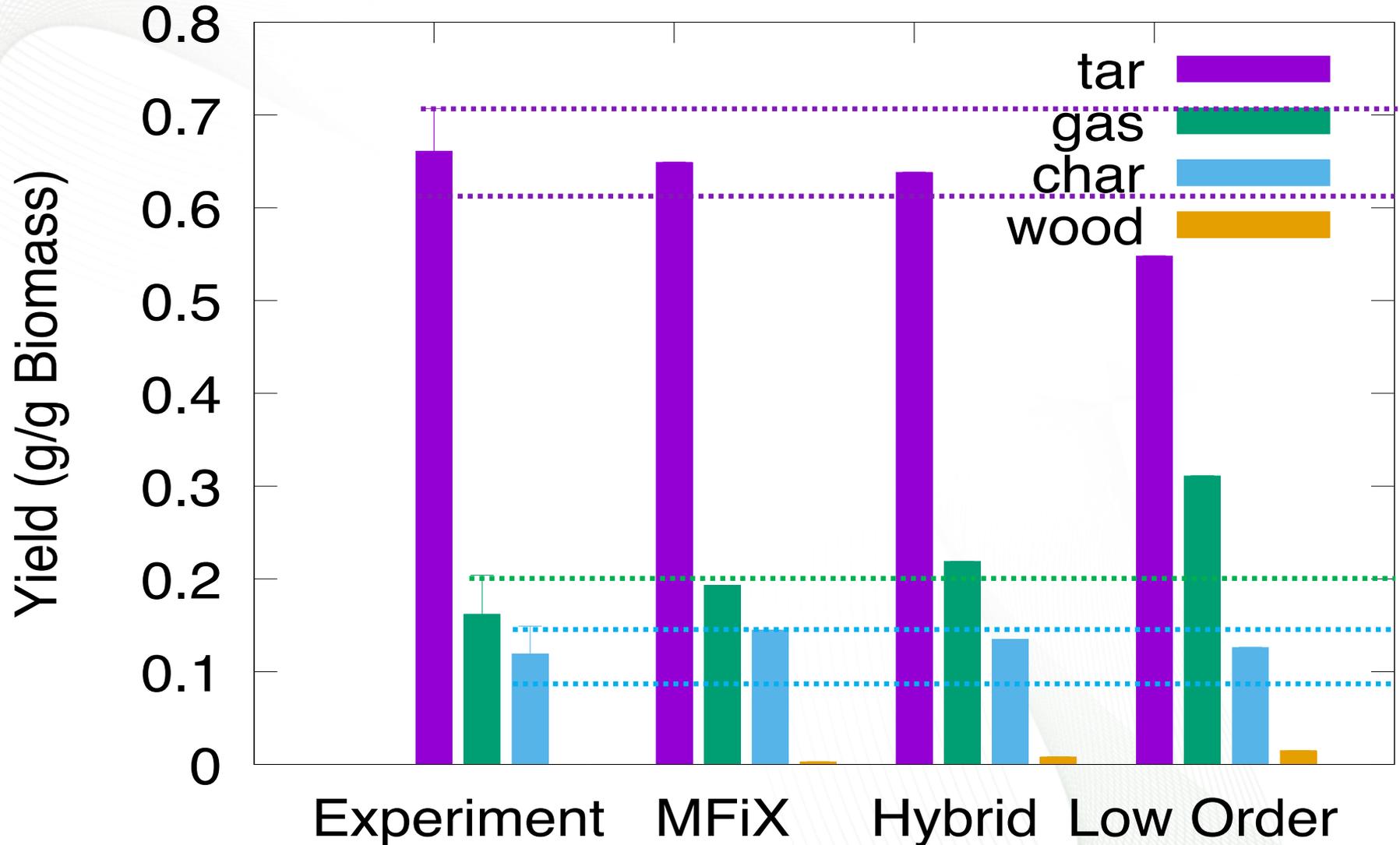
H.C. Park, H.S. Choi, The segregation characteristics of char in a fluidized bed with varying column shapes, Powder Technology 246 (2013) 561-571.

F. Berruti, A.G. Liden, D.S. Scott, Measuring and modelling residence time distribution of low density solids in a fluidized bed reactor of sand particles, Chemical Engineering Science 43 (1988) 739-748.

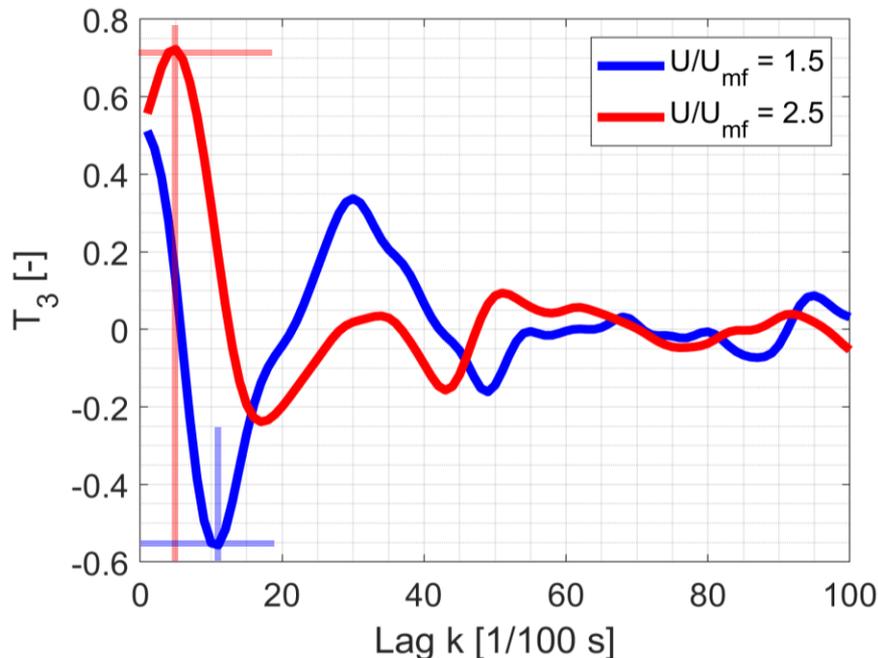
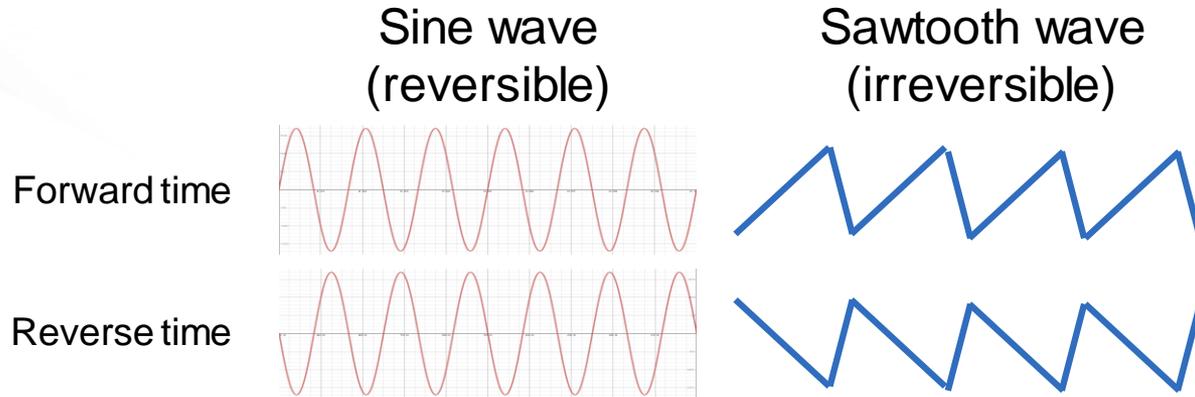
# NREL pyrolyzer conditions

Property	Units	Values
Particle diameter (sand)	$\mu\text{m}$	500
Particle density (sand)	$\text{kg}/\text{m}^3$	2500
Particle diameter (styrofoam/char)	$\mu\text{m}$	278
Particle density (styrofoam)	$\text{kg}/\text{m}^3$	-
Particle density (char)	$\text{kg}/\text{m}^3$	80
Temperature	K	773
Pressure (inlet)	kPa	133
Fluidizing $\text{N}_2$ (range)	m/s	0.13 – 0.47
Minimum fluidization (at 773 K)	m/s	0.0565
Coefficient of restitution	—	0.9
Angle of repose	$^\circ$	30
Friction coefficient	—	0.1

# Comparison of the three models with experimental yields



# Time-irreversibility functions show change in bubble passage profiles



$$T_3(k) = \sqrt{N-k} \frac{\sum_i^{N-k} (x_{i+k} - x_i)^3}{[\sum_i^{N-k} (x_{i+k} - x_i)^2]^{3/2}}$$

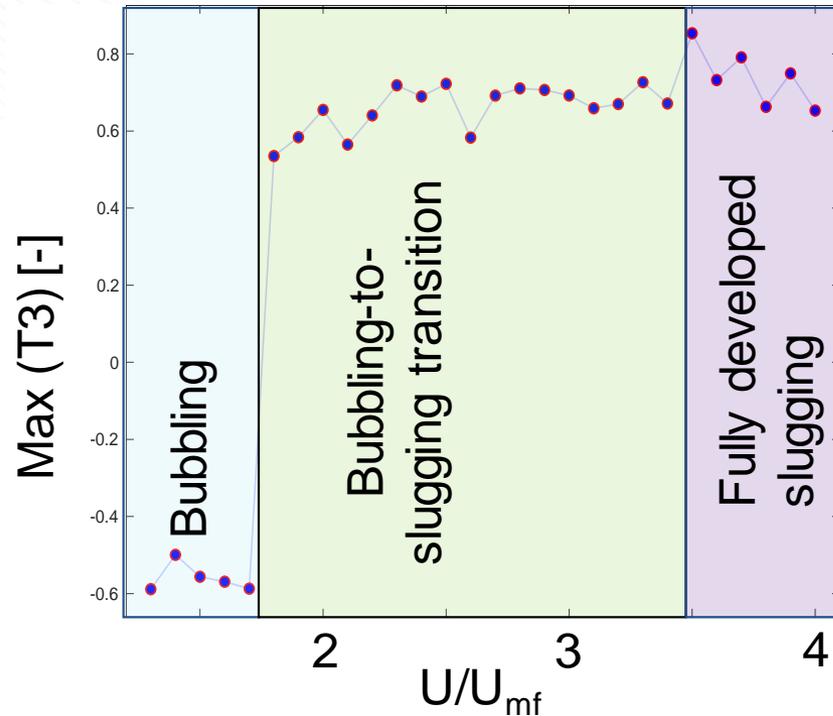
Summary metrics: magnitude and location (lag) of maximum  $T_3$

Observable: time series of a variable at a chosen location

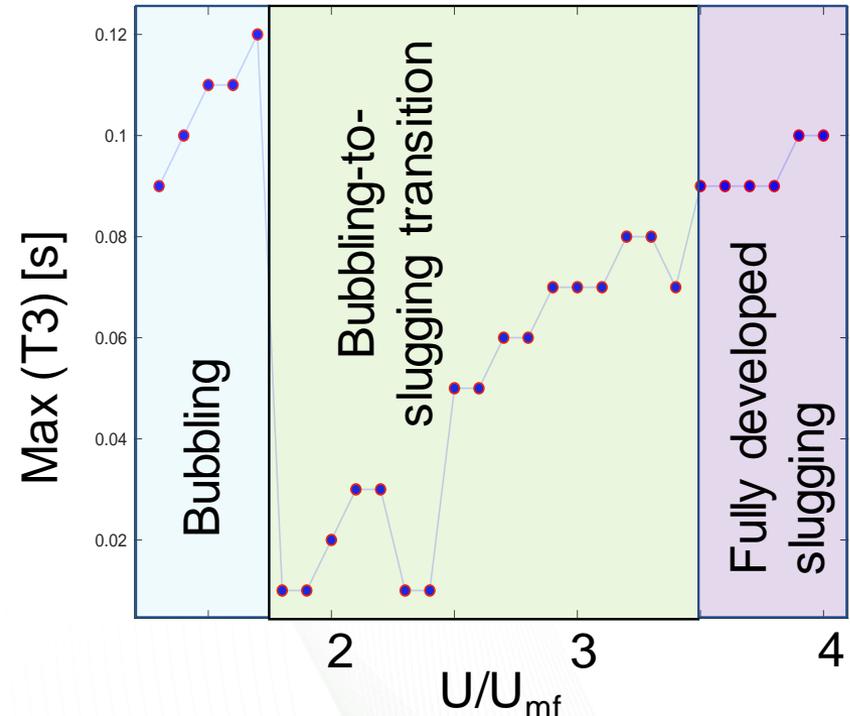
## Results (2)

# Time-irreversibility metric captures bubbling-to-slugging transition

Maximum time irreversibility



Location of maximum time irreversibility



- Observable: time series of bed pressure averaged over slice of  $0.9-1.0 \cdot H_0$
- “Jitter” due to finite-sample effects (limited observation time of  $\sim 30-40$  s)

## Results (3)

# Char mixing changes with fluidization regime

Fluidization regime affects char particle mixing intensity

Hydrodynamics must be considered

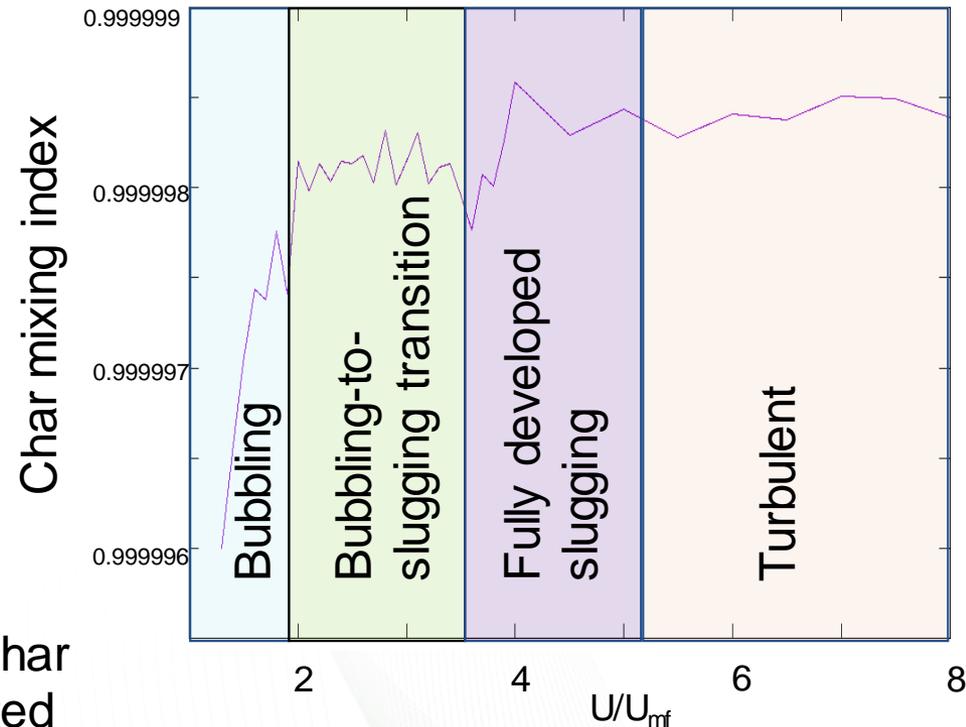
Kramer's mixing index

$$M = \frac{\sigma_0^2 - \sigma^2}{\sigma_0^2 - \sigma_r^2}$$

$$\sigma = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2}$$

$\sigma_0^2$  = standard deviation mass fraction of char when sand and char completely segregated

$\sigma_r^2$  = standard deviation mass fraction of char when sand and char completely mixed



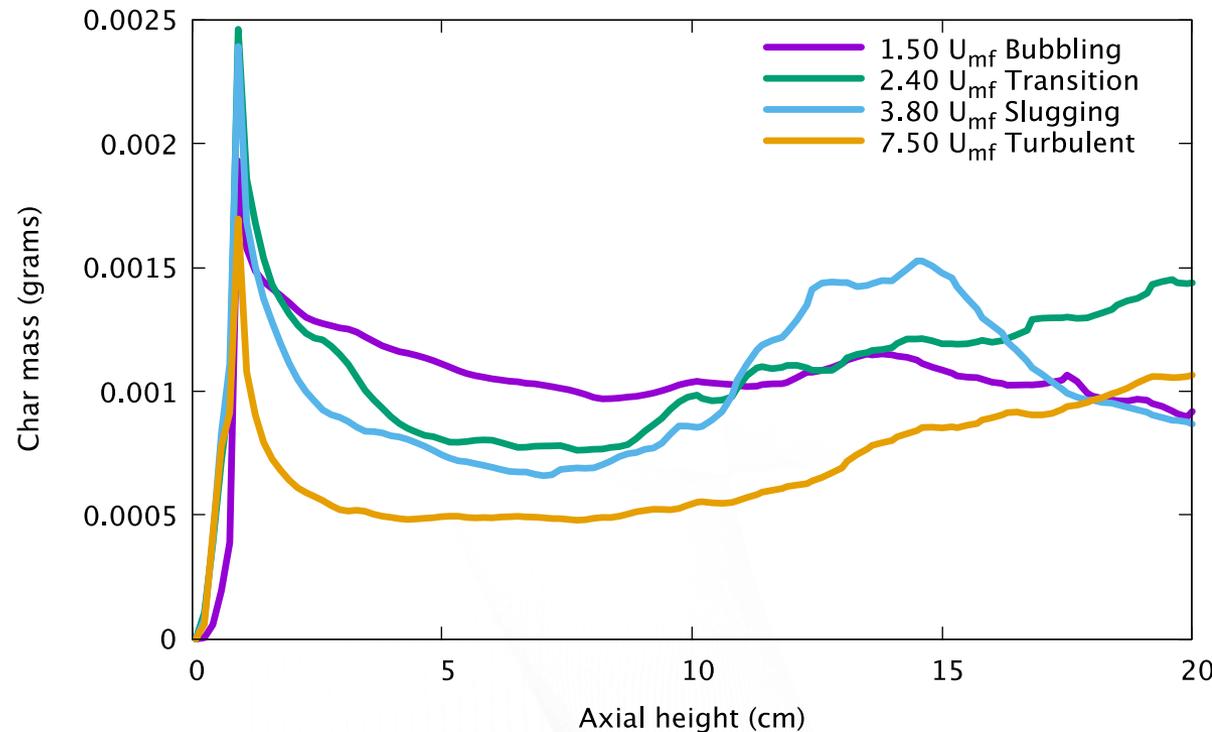
Gyenis, J. (1999). Assessment of mixing mechanism on the basis of concentration pattern. *Chemical Engineering and Processing: Process Intensification*, 38(4), 665-674. doi:10.1016/S0255-2701(99)00066-5

## Results (4)

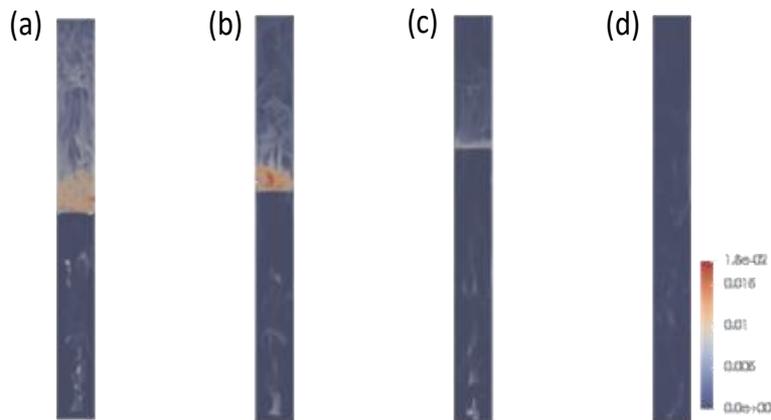
# Char spatial distribution changes with fluidization regime

- At bubbling-to-slugging transition char concentration at bottom decreases, but increase in top
- At slugging more char in top region
- At turbulent least char in the bed

Char mass vs axial height



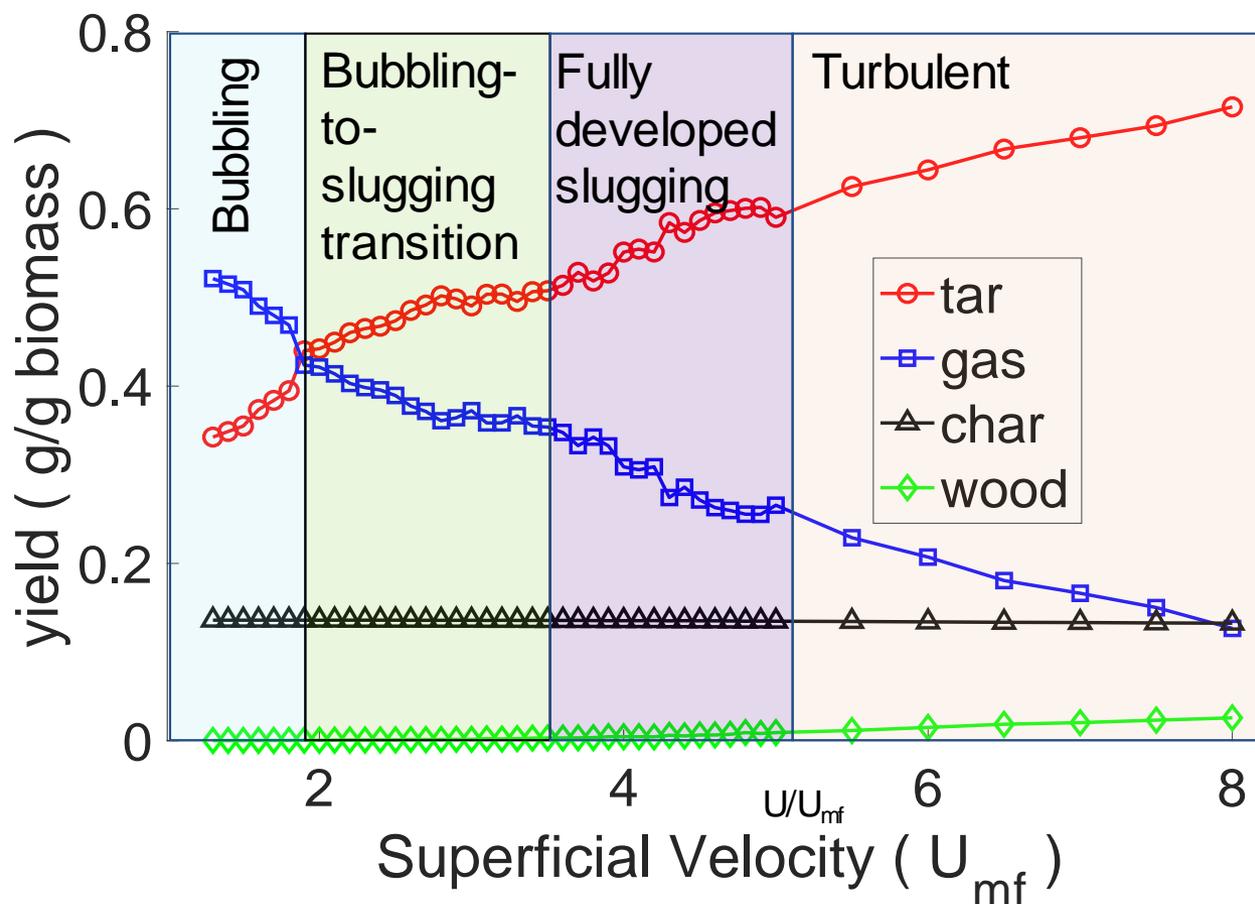
Note: this considers up to  $H_0$  and does not include freeboard.



# Bubbles/mixing/elutriation affect pyrolysis yield

Slugging beds reach a maximum in tar (oil) yield in the bubbling-to-slugging transition

Maximum tar (oil) yield occur at turbulent fluidization where slip velocity between gas and particles is high with a very short residence time

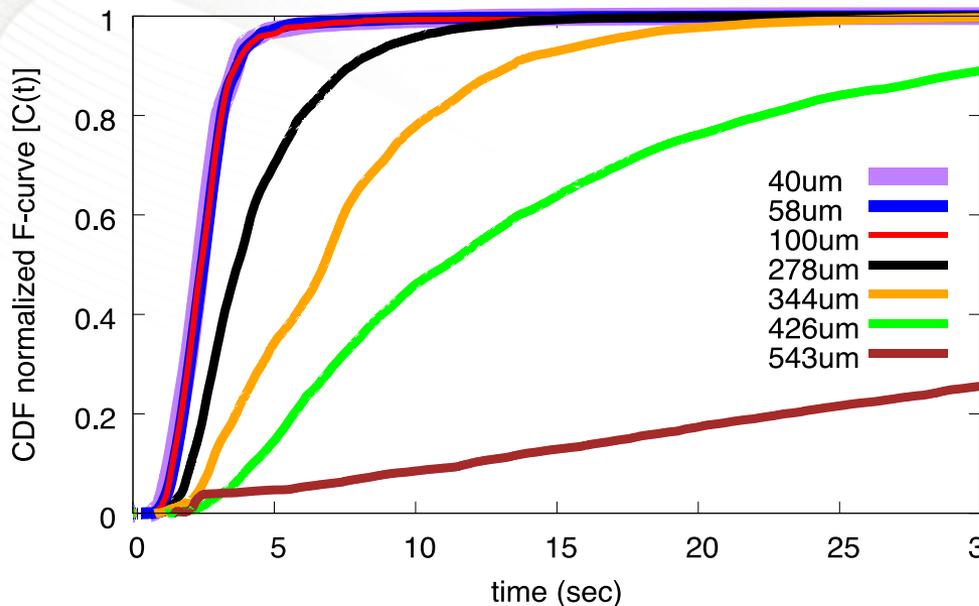


- “Jitter” due to finite-sample effects (limited bubble events seen during tracing)

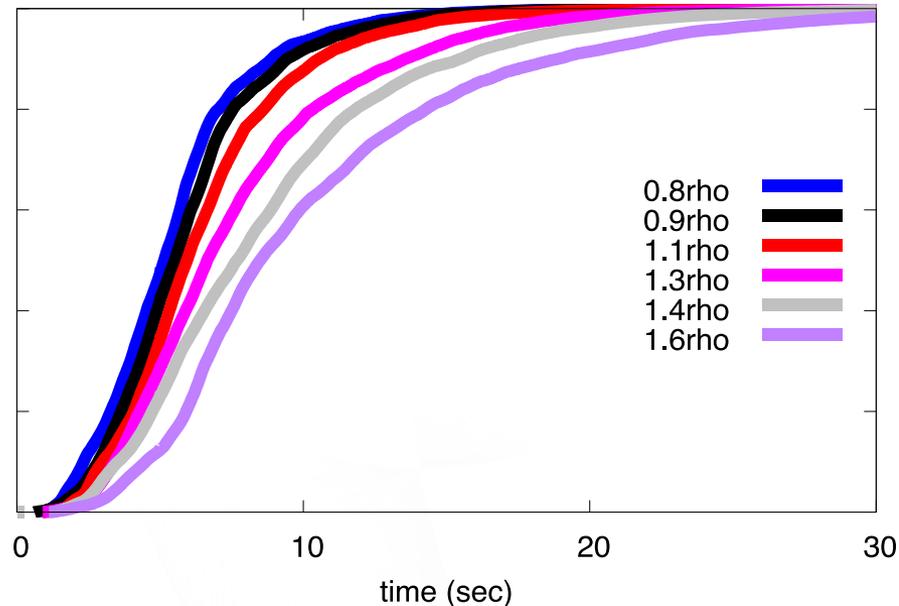
## Results (6)

# Particle size and density affect residence time distribution

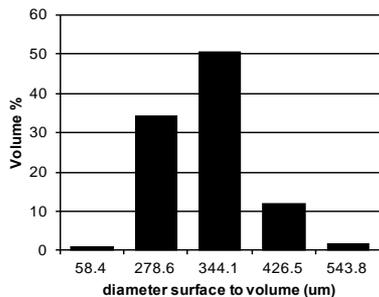
### Biomass particle size



### Biomass particle density



### Pine pellets milled 2mm

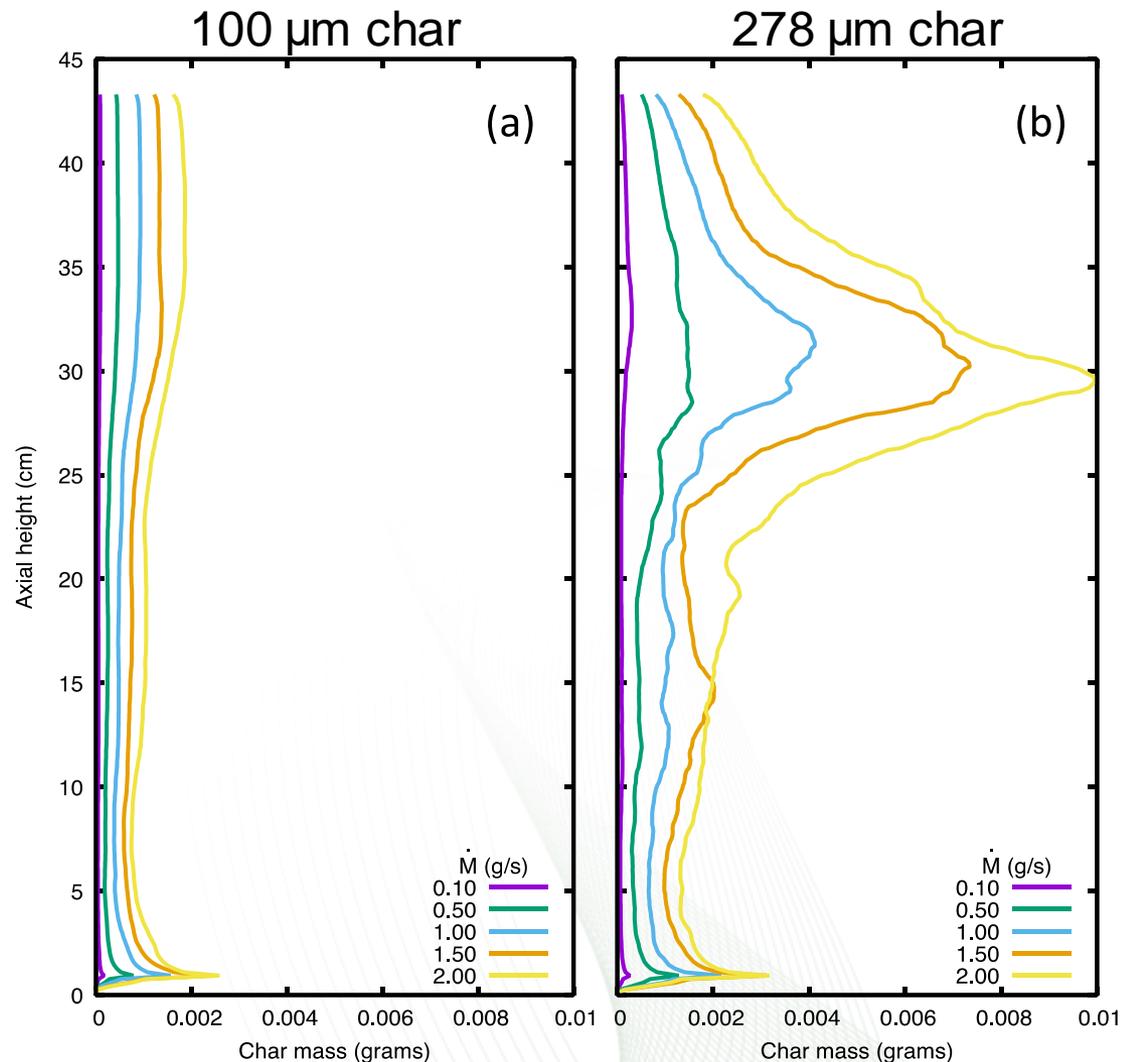


**Particle size and density must be selected carefully such that elutriation will occur**

Pecha, M. B., Ramirez, E., Wiggins, G. M., Carpenter, D., Kappes, B., Daw, S., & Ciesielski, P. N. (2018). "Integrated Particle- and Reactor-Scale Simulation of Pine Pyrolysis in a Fluidized Bed." *Energy & Fuels*, 32(10), 10683-10694.

# Char concentration varies with particle size and feed rate

- Larger particles create char layer at top of sand bed and freeboard region
- Tar vapors may be reduced through the char layer
- Char particle concentration changes bed particle size distribution and possibly fluidization
- 0.1181 g/s particle feed rate
- Monodisperse cases from among 100–500  $\mu\text{m}$  PSD



## Concluding remarks

- Quantifying the combined effects of hydrodynamics and chemistry is essential in utilizing lab-scale biomass pyrolysis reactor data for scale up
- Biomass particle properties and fluidization intensity have major impacts on product yields
- Two-fluid codes like MFiX can yield useful details about pyrolyzer hydrodynamics and gas and solid RTDs but improvements to the physics are still needed
- Combining MFiX hydrodynamics with low-order chemistry models appears to offer potential benefits
- Biomass pyrolysis reactor geometry and operating conditions must be designed in conjunction with a model that can capture the physics of interest
- **Biomass particle properties and feed rate have the potential to negatively affect pyrolysis yield and composition**
- **Char mixing and concentration in the bed and freeboard should be considered at the reactor design stage**

# Acknowledgements

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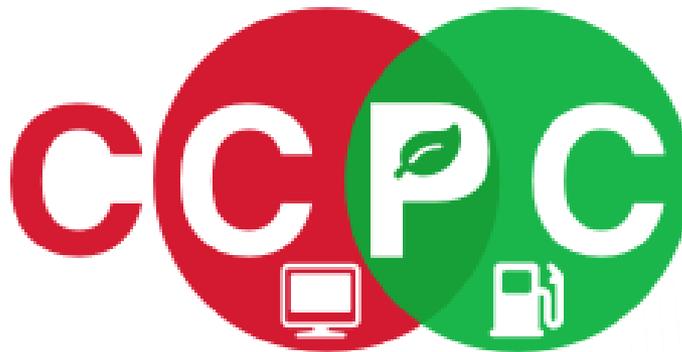
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# Questions?

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# Extra slides if there are questions

## Background and Motivation (3)

### How should lab FBR data be interpreted/analyzed?



FB Hydrodynamics directly impact:

1. Particle residence time
2. Gas residence time
3. Particle heating rate
4. Particle attrition/fragmentation
5. Particle and ash elutriation
6. Particle segregation

**All the above significantly impact raw oil yield and composition.**

Note: Bubble boundary depicted where void fraction  $> 0.65$

E. Ramirez, C.E.A. Finney, S. Pannala, C.S. Daw, J. Halow, Q. Xiong, Computational study of the bubbling-to-slugging transition in a laboratory-scale fluidized bed, *Chemical Engineering Journal* 308 (2017) 544-556. <http://dx.doi.org/10.1016/j.cej.2016.08.113>

# Kramer's mixing index for char mixing

- $M = \frac{\sigma_0^2 - \sigma^2}{\sigma_0^2 - \sigma_r^2}$  m=1 complete mixing, m=0 complete segregation

- $\sigma = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2}$

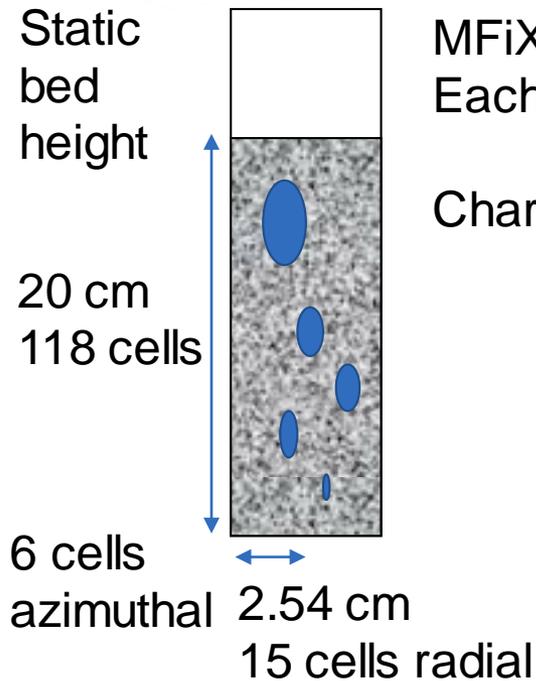
Created segregated and complete mixed case for analysis

- $\sigma_0^2$  = standard deviation mass fraction of char when sand and char completely segregated (heterogeneous)
- $\sigma_r^2$  = standard deviation mass fraction of char when sand and char completely mixed (homogeneous)

Guyot, J. (1999). Assessment of mixing frequency on the basis of concentration pattern. *Chemical Engineering and Processing: Process Intensification*, 38(4), 665-674. doi:10.1016/S0255-2701(99)00066-5

Pu, W., Zhao, C., Xiong, Y., Liang, C., Chen, X., Lu, P., & Fan, C. (2010). Numerical simulation on dense phase pneumatic conveying of pulverized coal in horizontal pipe at high pressure. *Chemical Engineering Science*, 65(8), 2500-2512. doi:10.1016/j.ces.2009.12.025

# Complete mixing case (Homogeneous)



MFiX has multiple cells 10620 cells in static bed height  
Each cell char and sand mass measured (time averaged time 15-19 s)

Char fraction in each cell used for stats (standard deviation)

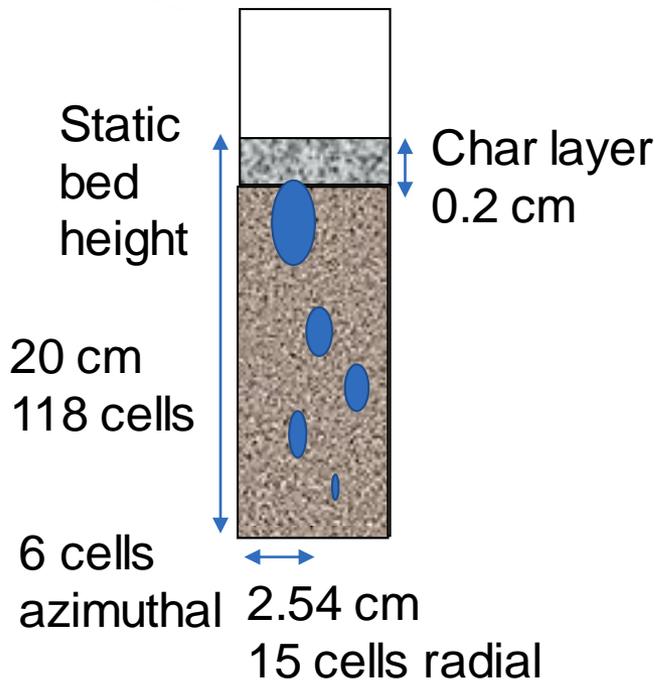
$$\sigma_r^2 = 0 \quad (\text{standard deviation})$$

Assumption for fully mixed:  
Bed region emulsion and bubbles  
have the same char fraction  
throughout

# Complete segregated case (Heterogenous)

MFiX has multiple cells 10620 cells in static bed height  
Each cell char and sand mass measured (time averaged time 15-19 s)

Char fraction in each cell used for stats (standard deviation)  
Char layer volume based on 0.51 void fraction (expanded bed-fluidized).



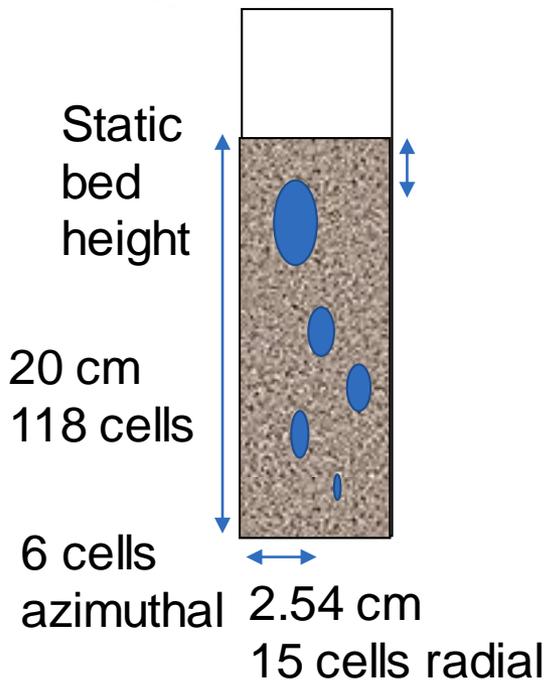
$$\sigma_0^2 = 0.312996541687 \quad (\text{standard deviation})$$

Assumption for fully segregated:  
Char layer above bed of sand is just char. Bubbles and emulsion in char layer have the same char fraction = 1.

# Mixing cases (1.3 – 8.0 $U_{mf}$ )

MFiX has multiple cells 10620 cells in static bed height  
Each cell char and sand mass measured (time averaged time 15-19 s)

Char fraction in each cell used for stats (standard deviation)



$\sigma_r^2$  (standard deviation)

Assumption for mixing cases:

Char mass fraction =  
char mass / (char mass + sand mass)

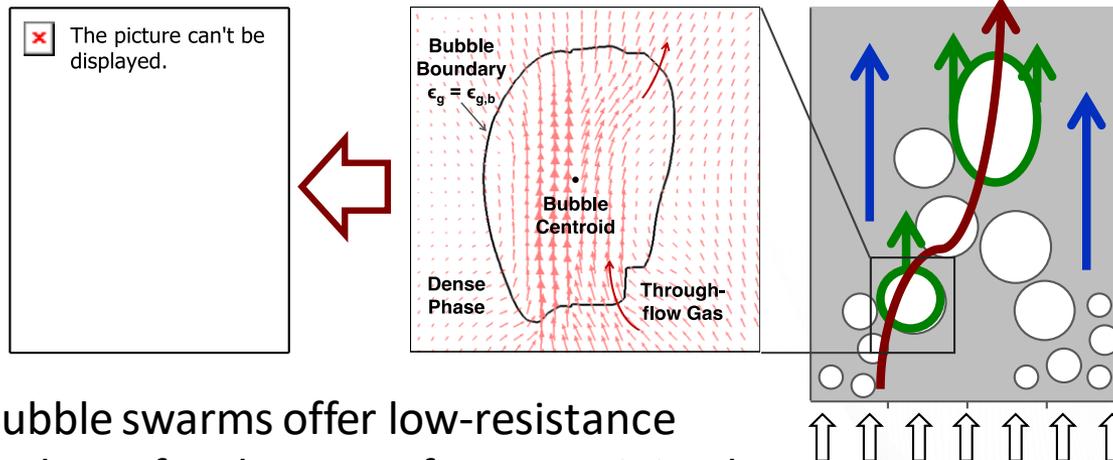
Averaged of 15.0 – 19.0 seconds

Bubbling bed at stationary state

# Phase 3: Preliminary Work

What is unique about bubbles that affects RTD and yields?

Total Gas Flow = dense flow + visible bubble flow + through-flow

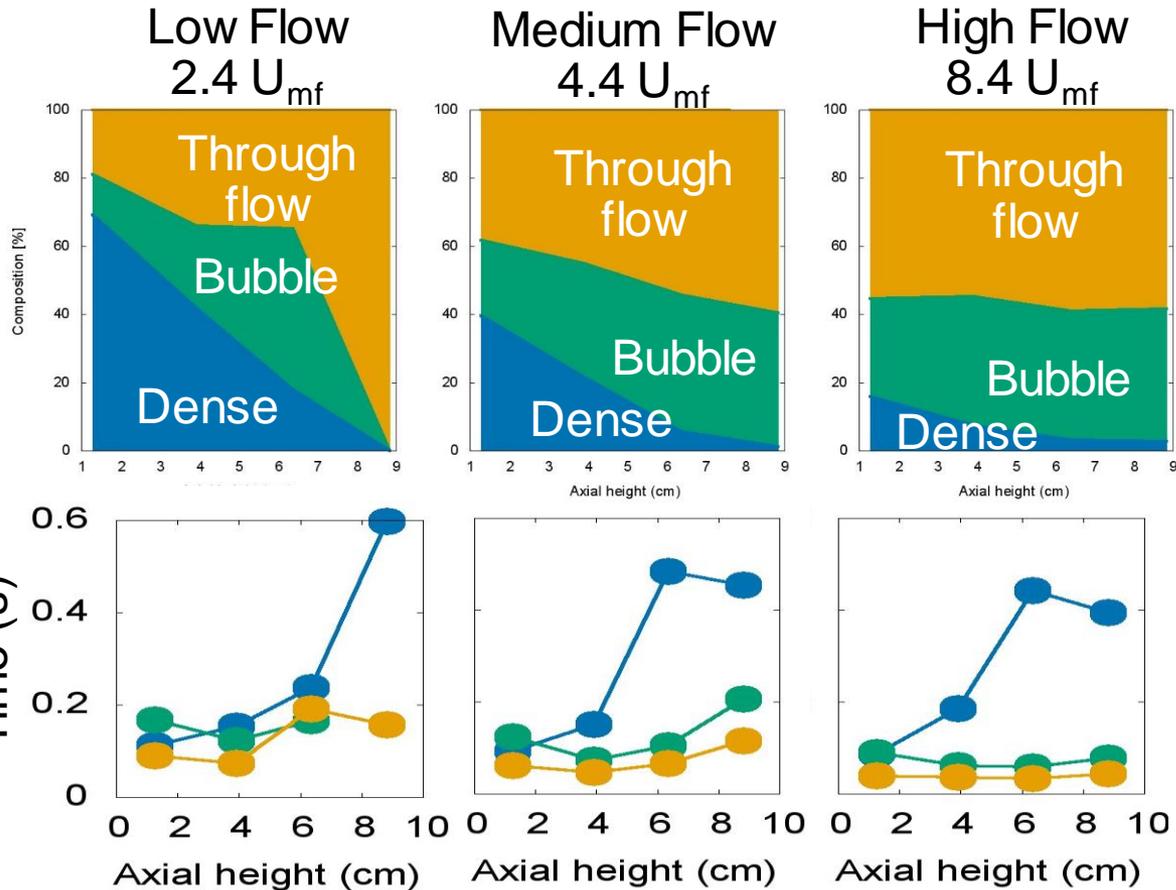
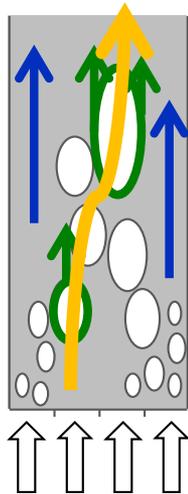


Bubble swarms offer low-resistance pathway for shortcut of gas => minimal contact with dense phase

A. Bakshi, C. Altantzis, R.B. Bates, A.F. Ghoniem, Multiphase-flow statistics using 3d detection and tracking algorithm (ms3data): Methodology and application to large-scale fluidized beds, Chemical Engineering Journal 293 (2016) 355-364. <http://dx.doi.org/10.1016/j.cej.2016.02.058>

# Phase 3: Preliminary: Pyrolysis chemistry + CFD

Bubbles affect residence time along reactor height



# Peer reviewed publications

Ramirez, E., Li, T., Shahnam, M., & Daw, C. S. (**In Preparation**). "Computational study on biomass fast pyrolysis: Hydrodynamic effects on the performance of a laboratory-scale fluidized bed reactor." Chemical Engineering Journal.

Ramirez, E., Finney, C.E.A., Daw, C. S. (**In Preparation**). "Computational study on biomass fast pyrolysis: Design considerations for a laboratory-scale fluidized bed." Chemical Engineering Journal.

Pecha, M. B., Ramirez, E., Wiggins, G. M., Carpenter, D., Kappes, B., Daw, S., & Ciesielski, P. N. (2018). "Integrated Particle- and Reactor-Scale Simulation of Pine Pyrolysis in a Fluidized Bed." Energy & Fuels, 32(10), 10683-10694.

Ramirez, E., Finney, C.E.A., Pannala, S., Daw, C.S., Halow, J., Xiong, Q. (2017). "Computational study of the bubbling-to-slugging transition in a laboratory-scale fluidized bed." Chemical Engineering Journal, **308**: 544-556.

Xiong, Q., et al. (2016). "Modeling the impact of bubbling bed hydrodynamics on tar yield and its fluctuations during biomass fast pyrolysis." Fuel **164**: 11-17.

Daw, C.S., Wiggins, G., Xiong, Q., Ramirez, E. (2016) "Development of a Low-Order Computational Model for Biomass Fast Pyrolysis: Accounting for Particle Residence Time." ORNL/TM-2016/69.

Clark, E., Griffard, C., Ramirez, E., & Ruggles, A. (2015). Experiment attributes to establish tube with twisted tape insert performance cooling plasma facing components. Fusion Engineering and Design, **100**, 541-549.